

INSAR PERFORMANCE FOR LARGE-SCALE DEFORMATION MEASUREMENT: IMPACT OF TROPOSPHERIC CORRECTIONS AND VALIDATIONS

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ABSTRACT

This paper deals with the analysis of InSAR performance for large-scale deformation measurement. The study evaluates the use of models, especially numerical weather prediction re-analysis, to mitigate disturbances in SAR interferograms. The impact of such corrections is evaluated and, using GNSS measurements, the predicted error bars are validated on a large Sentinel-1 dataset.

Index Terms— InSAR, performance, deformation, numerical weather prediction (NWP), GNSS

1. A MODEL FOR TROPOSPHERE DRIVEN INTERFEROMETRIC PERFORMANCE

The use of NWP for mitigating the tropospheric signal in the interferometric phase is limited both in accuracy, due to the limitations of the model itself, and in resolution. The best sampling available is 9 km in the case of ECMWF operational products, several orders of magnitude larger than the SAR resolution, even after interferometric multi-looking. The accuracy of NWP can be strongly site dependent since it depends on data availability and the fit of the model to actual tropospheric conditions. For the ECMWF products considered here, there is a strong dependence of accuracy on latitude [1]. However, it should be pointed out that since the interferograms are intrinsically relative, the impact of model bias in the simulated interferometric phase is limited. More important is the variance of the error between the real zenith delay z_t and the modeled delay \hat{z}_t $\epsilon = z_t - \hat{z}_t$. This can be seen as a Gaussian process in time and space with mean μ_ϵ , variance σ_ϵ^2 and spatial correlation function Λ . The interferometric error between two points A and B for reference acquisition r and secondary acquisition s is

$$\phi_\epsilon = (\epsilon_{r,A} - \epsilon_{s,A}) \frac{e^{-\frac{h_A}{H}}}{\cos \theta_A} - (\epsilon_{r,B} - \epsilon_{s,B}) \frac{e^{-\frac{h_B}{H}}}{\cos \theta_B} \quad (1)$$

where θ is the local incidence angle. The effect of the altered path delay due to topography is taken in account using the exponential mapping function as in [2] with h_A and

h_B the height of the points w.r.t to the ellipsoid and H , the so called height of the atmosphere, typically fixed to 7 km. Assuming the error is first-order stationary, the mean interferometric phase error is $\mathbb{E}[\phi_\epsilon] = 0$ even if the residual error has non-zero mean, $\mu_\epsilon \neq 0$ as visible from Equation 1. Assuming the atmosphere and the error are temporally uncorrelated, the interferometric phase error power is

$$\mathbb{E}[\phi_\epsilon^2] = 2\sigma_\epsilon^2 \frac{e^{-\frac{2h_B}{H}} \cos^2 \theta_A + e^{-\frac{2h_A}{H}} \cos^2 \theta_B}{\cos^2 \theta_A \cos^2 \theta_B} - \frac{4\Lambda(A, B)}{\cos \theta_A \cos \theta_B} e^{-\frac{h_B + h_A}{H}} \quad (2)$$

Λ is a correlation function that $\Lambda(A \equiv B) = \sigma_\epsilon^2$ and $\Lambda(A, B) \rightarrow 0$ at large distance. The residual error in the interferometric phase between points A and B described in Equation 2 is a parametrization of the sample variograms that can be calculated from data. Equation 2 approaches 0 as the distance $d \rightarrow 0$ and is $\propto \sigma_\epsilon^2$ as $d \rightarrow \infty$. That means that at large distances (≥ 40 -50 km) the interferometric measurements become limited by the accuracy of the models used for the correction.

In order to study the troposphere induced error and its mitigation through the use of NWP, short time interferograms are generated from every acquisition with the next. This approach should prevent bias due to a common reference as well as eventual seasonality present in the tropospheric delay and the impact of deformation. The latter point requires clarification. The fast revisit time of the Sentinel-1 mission allows temporal baselines of 6,12 or in worse cases, 24 days. Since the residual error, including effects such as tropospheric turbulence, are assessed to be on the order of centimeters, deformation rates of several tens of cm/y would be necessary in order to produce effects comparable to those from the troposphere.

The tropospheric error is considered temporally uncorrelated. This hypothesis should hold at least for the troposphere corrected phase since all seasonal effects are included in the NWP and are hence compensated. The interferometric phase variogram between acquisitions n and m , $\Gamma_{n,m}$, is then the sum of the two variograms of the errors in acquisition n and

m , Γ_n and Γ_m

$$\begin{aligned}\Gamma_{n,m}(d) &= \mathbb{E}\left[\left((\phi_n(A) - \phi_n(B)) - (\phi_m(A) - \phi_m(B))\right)^2\right] \\ &= \Gamma_n + \Gamma_m\end{aligned}\quad (3)$$

where ϕ is the phase at points A and B and d is the distance between the points. By averaging the set of short time baseline variograms $\bar{\Gamma} = \mathbb{E}[\Gamma]$, it is possible to derive the average behavior of the residual interferometric phase error. Performing this experiment before and after the NWP correction allows an analysis of the impact of the corrections at different scales. The interferometric data and NWP models have different resolutions, hence different behavior is expected at different scales. This is particularly important in tectonic strain applications where high relative accuracy is required at large scales. Assuming the troposphere is the main source of error at large distances (≥ 40 -50 km), one can now derive its impact on deformation rate measurements [3]. The variogram of the linear deformation rate estimates is obtained by scaling $\bar{\Gamma}$ by the linear regression formula [4],

$$\Gamma_v(d) = \frac{1}{2} \frac{\lambda^2}{16\pi^2} \frac{\bar{\Gamma}(d)}{M\sigma_t^2} \quad (4)$$

where the factor $1/2$ accounts for the common reference acquisition, M is the number of acquisitions and σ_t^2 is the spread of the temporal sampling $\sigma_t^2 = \sum_i t_i^2 / M - (\sum_i t_i / M)^2$

2. IMPACT OF THE TROPOSPHERIC CORRECTIONS

Variograms are a convenient way to characterize the tropospheric component of the deformation rate precision. By comparing the interferometric phase variograms before and after the correction, it should be possible to observe the gain at different scales. Since the data used for compensation are not provided at the same resolution of the interferograms, a significant reduction of the tropospheric contribution can only be expected at scales larger than the NWP model resolution. Here, ECMWF ERA-5 data are used with a spatial gridding of 30 km. It should be noted that the accuracy of the correction has been demonstrated to vary considerably worldwide [1].

In order to comprehensively evaluate correction performance, 146 Sentinel-1 stacks have been processed over various regions of the world. Each of these stacks spans at least 4 years and contains up to 200 acquisitions. As described in Section 1, only short time baseline interferograms were used to ensure the troposphere signal dominates.

An evaluation of the performance at large distances was also performed. The average variograms before and after corrections were fitted using typical variogram models (exponential, Gaussian and spherical) hence reducing the variance

of the variogram estimate. Using this information together with temporal sampling information and Equation 4, the performance in measuring the deformation rates at 150 km was estimated with and without the tropospheric corrections.

Mitigation of tropospheric effects and the correct characterization of their spatial characteristics are critical in developing error models for missions that allow [5] or will allow [6] systematic interferometric processing on a global scale.

Figure 1 show a significant gain is attained when the tropospheric delay is corrected. As already pointed out in [1] the gain is region dependent. A very high gain is achieved where the model accuracy is also very good such as in Europe or the U.S. In only one of the 146 stacks, located in Indonesia, were the corrections found to slightly worsen the performance.

3. ERROR-BARS VALIDATION

Knowledge of displacement rate precision is of central importance in geophysical studies of large-scale tectonic movements where the precision requirements are often very high [7]. For the sake of completeness, the behavior of the displacement rate as described in Equation 4 must be verified in practice. This was achieved by cross-validation with GNSS measurements.

Given a full statistical characterization of the GNSS/InSAR differences [8][9], and assuming the GNSS accuracies to be correct, we can now verify if Γ_v actually represents the variogram of the velocity error. First, from the vector $\underline{\Delta}$ of N differences between GNSS and InSAR velocities, we compute all $(N^2 - N)/2$ unique pairwise differences between the elements. Their mean is expected to be zero, with $\sigma_{G,i}^2$ the error variance of the i -th GNSS measurement, their variance is

$$\sigma_{\Delta}^2(i, j) = \mathbb{E}[(\Delta_i - \Delta_j)^2] = \sigma_{G,i}^2 + \sigma_{G,j}^2 + \Gamma_v(d_{i,j}) \quad (5)$$

The scales at which Γ_v can be verified by this method clearly depend on the spatial distribution of GNSS stations, with the range of scales being limited by the smallest and largest distances between stations. A statistic defined as the standardized difference

$$T_{i,j} = \frac{\Delta_i - \Delta_j}{\sigma_{\Delta}(i, j)} \quad (6)$$

should follow a standard normal distribution if the error description is correct. To test the validity of the variance model in Equation 5, a χ^2 -test can be performed on \underline{T} . The confidence region at significance level α for the true σ_T^2 is

$$\sqrt{\frac{(N-1)\hat{\sigma}_T^2}{\chi_{1-\alpha/2, N-1}^2}} < \sigma_T < \sqrt{\frac{(N-1)\hat{\sigma}_T^2}{\chi_{\alpha/2, N-1}^2}} \quad (7)$$

where the number of degrees of freedom is $N - 1$ and $\hat{\sigma}_T^2$ is the sample variance of T . The validation approach will

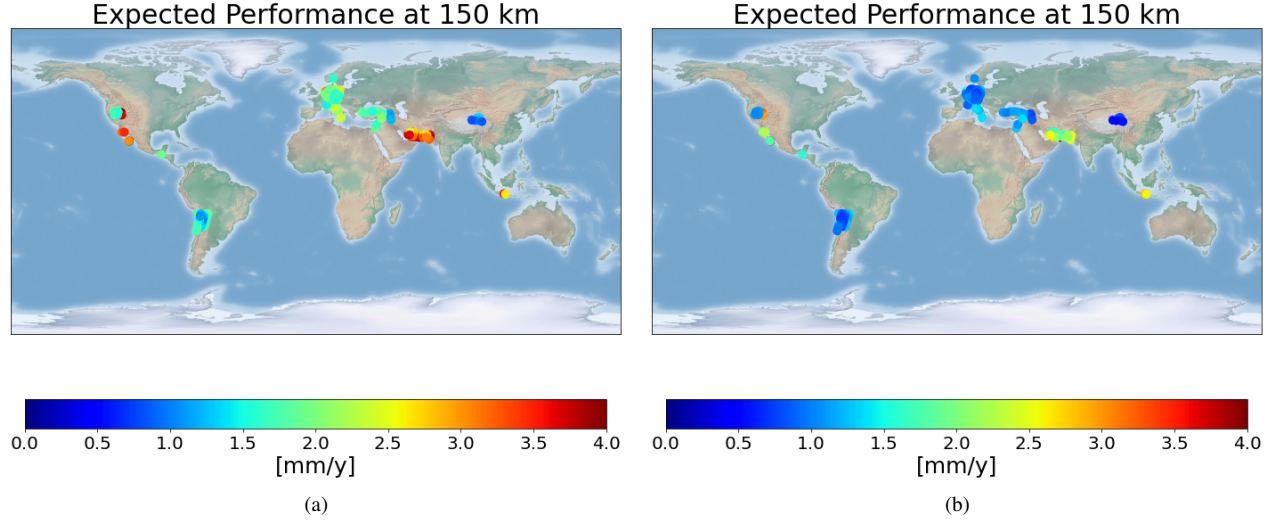


Fig. 1: Geographic distribution of the expected performance at 150 km: before (a) and after (b) the corrections..

focus on analyzing the distribution of \underline{T} and the confidence interval for σ_T . The cross-validation scheme was performed on the German deformation map. The dataset includes 41 stacks acquired in both ascending and descending geometry. This region was chosen due to easy and open access to a very dense GNSS network from Nevada Geodetic Laboratories [10, 11] with all stacks containing sufficient GNSS stations.

The confidence intervals for σ_T at 5% significance level are shown in Figure 2 (a). Of course, the dependence on the number of GNSS stations, is strong. However, for almost all stacks the confidence intervals include $\sigma_T = 1$ indicating acceptance of the null hypothesis $H_0 : \sigma_T = 1$ versus the alternative $H_1 : \sigma_T \neq 1$.

Figure 2 (b) shows the distribution of \underline{T} pooled over all stacks, along with the nominal standard normal distribution under the null. The pooled value of σ_T is 1.03, showing very good agreement between the model and the measurements.

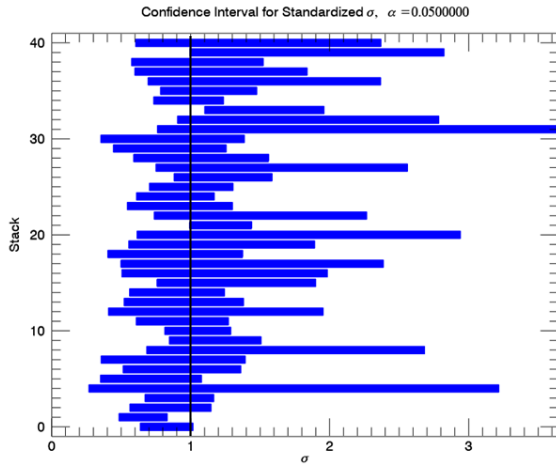
4. CONSIDERATIONS ABOUT THE DETECTION OF NON-LINEAR DEFORMATION

To conclude it is worthwhile consider the capability of InSAR measuring the deviation of the time series from the model used for the temporal phase unwrapping. These signals can be very interesting for scientists and are often observed in tectonics motions (i.e. post-seismic deformation, slow slip events). While the achievable accuracy of the deformation rates takes advantage from a longer time series (Equation 4) this is not true when retrieving the non-linear part of the deformation signal that is bound to the accuracy of a single phase sample as indicated by $\bar{\Gamma}(d)/2$ in Equation 4. According to the data collected in this study, the accuracy of the single phase sample is on the order of centimeters (depending on the geographic area and time of day). The detectability of transient signals is

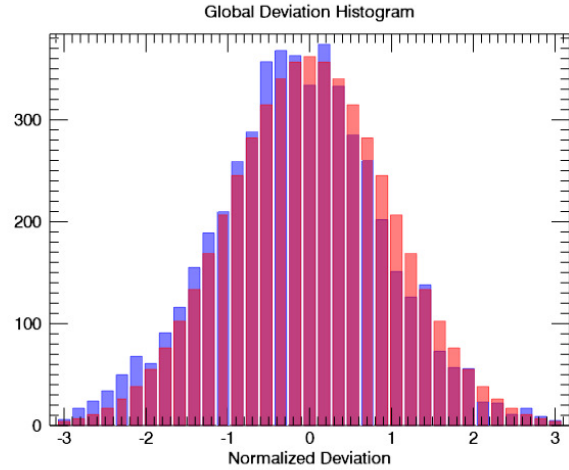
determined by intensity and scale. Figure 3 shows the average of the variograms computed on short temporal baselines of the residual phase obtained compensating the linear model estimated with the temporal phase unwrapping. This should represent the accuracy in detecting transient signals at different scales. On scales greater than 100 km, only several centimeters of non-linear deformations are detectable in InSAR time series while at medium scales (40-50 km) smaller event such like volcanic unrest are observable. Of course assumptions on the deformation models can improve this capability. Spatial/temporal filtering can improve the precision by trading it off against a reduced observability of the temporal/spatial scale of the deformation signal.

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(a)



(b)

Fig. 2: Results of the InSAR/GNSS velocities cross-validation: (a) the confidence intervals where the black line denotes $\sigma_T = 1$. (b) Histogram of the pooled vectors \underline{T} using NGL GNSS results (blue) compared to the nominal $N(0, 1)$ distribution (red).

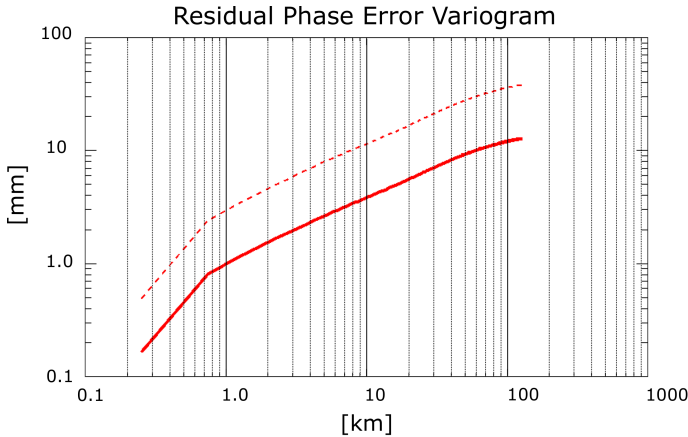


Fig. 3: Residual phase variogram 1σ (bold line), 3σ (dashed line)

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